```
model {
### Model for ipd ###
for(i in 1:n.subjects.ipd) {
      #likelihood
      y.ipd[i] ~ dnorm(theta[i], prec.ipd[study.ipd[i]])
      #model
      theta[i] <- mu.ipd[study.ipd[i]] +</pre>
                                               b_basey[study.ipd[i]] *
      basey.ipd[i] +
      b_age.ipd[study.ipd[i]]*age.ipd[i]*mod1+b_bmi.ipd[study.ipd[i]]*bmi.
       ipd[i]*mod2 +
       (delta[index.ipd[i]] +
(b_tx_age*age.ipd[i]*mod1+b_tx_bmi*bmi.ipd[i]*mod2)*
       equals(baseline.ipd[i],1))*(1 -equals(treat.ipd[i],baseline.ipd[i]))
      #residual deviance
      dev.ipd[i]<- (y.ipd[i] - theta[i]) * (y.ipd[i] - theta[i])</pre>
      #impute missing data
       age.ipd[i]~dnorm(mu.age.ipd,p.age.ipd)
      bmi.ipd[i]~dnorm(mu.bmi.ipd,p.bmi.ipd)
}
for(l in 1:(n.arms.ipd)) {
      #consistency model
      delta[1]~dnorm(md[1], prec.re.ipd[1])
      md[1]<- d[treat.ipd.sum[1]] - d[baseline.ipd.sum[1]] + sw.ipd[1]</pre>
      prec.re.ipd[1]<- prec.re * 2 * (m.ipd.sum[1] - 1)/m.ipd.sum[1]</pre>
      #residual deviance
       resdev.ipd[1]<- sum(dev.ipd[start.ipd.sum[1]:end.ipd.sum[1]]) *</pre>
prec.ipd[study.ipd.sum[1]]
}
      #multi-arm adjustment for up to 4 arm studies
       sw.ipd[1]<-0
       sw.ipd[2]<-0
for (l in 3:n.arms.ipd) {
       sw.ipd[1]<- ((delta[1 - 1] - d[treat.ipd.sum[1 - 1]] +</pre>
      d[baseline.ipd.sum[1 - 1]]) *
       (equals(m.ipd.sum[1],3)+equals(m.ipd.sum[1],4)) +
       (delta[1 - 2] - d[treat.ipd.sum[1 - 2]] + d[baseline.ipd.sum[1 -
      2]]) * equals(m.ipd.sum[1],4))/ (m.ipd.sum[1]-1)
}
      #priors for missing data imputation parameters
      mu.age.ipd~dunif(-50,50)
       sd.age.ipd~dunif(0,30)
      p.age.ipd<-1/pow(sd.age.ipd,2)</pre>
```

mu.bmi.ipd~dunif(-40,40)

```
sd.bmi.ipd~dunif(0,20)
      p.bmi.ipd<-1/pow(sd.bmi.ipd,2)</pre>
for(j in 1:n.trials.ipd) {
      #random effect on baseline outcome coefficient
      b basey[j]~dnorm(b basey mean,prec basey)
      #prior on covariate main effect coefficients, study-specific
baseline term, outcome se
      b_age.ipd[j]~dnorm(0,1.0E-4)
      b_bmi.ipd[j]~dnorm(0,1.0E-4)
      mu.ipd[j]~dnorm(0,1.0E-4)
      se.ipd[j]~dunif(0,2)
      var.ipd[j]<-se.ipd[j]*se.ipd[j]</pre>
      prec.ipd[j]<-1/var.ipd[j]</pre>
}
### Model for aggregate data ###
      #impute standard deviation and remove feedback
      sdev.imp~dgamma(alpha,beta)
       sdev.imp.cut<-cut(sdev.imp)</pre>
      #parameter to adjust for baseline imbalance, remove feedback
      b_basey_new~dnorm(b_basey_mean,prec_basey)
      ancova.cut<-cut(b_basey_new)</pre>
for (j in 1:n.arms.ad) {
      #standard deviation imputed for study arm
      sdev[j]<- sd.y.ad[j]*(1-equals(sd.y.ad[j],0)) +</pre>
sdev.imp.cut*equals(sd.y.ad[j],0)
      se[j]<-sdev[j]/sqrt(n.ad[j])</pre>
      prec.ad[j]<-1/(se[j]*se[j])</pre>
      #likelihood
      y.ad[j] ~ dnorm(theta.ad[j], prec.ad[j])
      #model
      theta.ad[j] <- mu.ad[study.ad[j]] +</pre>
              (delta[j + n.arms.ipd]+ancova.cut*imb.ad[j] +
                            (b_tx_age*age.ad[study.ad[j]]*mod1 +
      b_tx_bmi*bmi.ad[study.ad[j]]*mod2)*
              (equals(baseline.ad[j],1))) *(1-
      equals(treat.ad[j],baseline.ad[j]))
      #consistency model
      delta[j + n.arms.ipd] ~ dnorm(md.ad[j], prec.re.ad[j])
      md.ad[j]<- d[treat.ad[j]] - d[baseline.ad[j]] + sw.ad[j]</pre>
      prec.re.ad[j]<- prec.re * 2 * (m.ad[j] - 1)/m.ad[j]</pre>
      #residual deviance
```

```
resdev.ad[j] <- ((y.ad[j] - theta.ad[j]) * (y.ad[j] - theta.ad[j]) *</pre>
prec.ad[j])
}
      #multi-arm adjustment for up to 5 arm studies
       sw.ad[1]<-0
      sw.ad[2]<-0
for (i in 3:3) {
       sw.ad[i]<- ((delta[(i+n.arms.ipd) -1] - d[treat.ad[i-1]] +</pre>
       d[baseline.ad[i - 1]]) * (equals(m.ad[i],3)+equals(m.ad[i],4)) +
       (delta[(i + n.arms.ipd) - 2] - d[treat.ad[i - 2]] + d[baseline.ad[i
       - 2]]) * (equals(m.ad[i],4))) / (m.ad[i]-1)
}
for (i in 4:n.arms.ad) {
       sw.ad[i]<- ((delta[(i + n.arms.ipd) - 1] - d[treat.ad[i - 1]] +</pre>
       d[baseline.ad[i - 1]]) *
       (equals(m.ad[i],3)+equals(m.ad[i],4)+equals(m.ad[i],5)) +
       (delta[(i + n.arms.ipd) - 2] - d[treat.ad[i - 2]] + d[baseline.ad[i
       - 2]]) * (equals(m.ad[i],4)+equals(m.ad[i],5))+(delta[(i +
       n.arms.ipd) - 3] - d[treat.ad[i - 3]] + d[baseline.ad[i - 3]]) *
       (equals(m.ad[i],5))) / (m.ad[i]-1)
}
      #priors for missing data imputation parameters
      mu.age.ad~dunif(-50,50)
       sd.age.ad~dunif(0,30)
       p.age.ad<-1/pow(sd.age.ad,2)</pre>
      mu.bmi.ad~dunif(-40,40)
       sd.bmi.ad~dunif(0,20)
       p.bmi.ad<-1/pow(sd.bmi.ad,2)</pre>
for(j in 1:n.trials.ad) {
      #prior on study-specific baseline term
      mu.ad[j]~dnorm(0,1.0E-4)
      #impute missing data
       age.ad[j]~dnorm(mu.age.ad,p.age.ad)
       bmi.ad[j]~dnorm(mu.bmi.ad,p.bmi.ad)
}
      #sum of residual deviance for aggregate and individual patient data
      totresdev.ad<- sum(resdev.ad[])</pre>
      totresdev.ipd<- sum(resdev.ipd[])</pre>
###Priors for shared parameters###
      #priors on treatment effects
       d[1]<-0
for (k in 2:n.treatments) {
```

```
d[k] ~ dnorm(0,1.0E-3)
}
#prior on random treatment effect
tau~ dunif(0,2)
tau.sq <- tau *
prec.re<- 1 / tau.sq</pre>
```

#prior on mean and random effect variance for coefficient on baseline outcome

```
b_basey_mean~dnorm(0,1.0E-6)
tau_basey~dunif(0,2)
tau_basey.sq<-tau_basey*tau_basey
prec_basey<-1 / tau_basey.sq</pre>
```

```
#prior on covariate-treatment interaction coefficients
b_tx_age~dnorm(0,1.0E-4)
b tx bmi~dnorm(0,1.0E-4)
```

```
###Additional statistical summaries
# pairwise comparisons
for (c in 1: n.treatments) {
    for (k in 1: n.treatments) {
        ef[c,k] <- d[c] - d[k]
        }
}</pre>
```

```
# ranking and prob{treatment k is the best}
for (k in 1: n.treatments) {
        rk[k]<- n.treatments + 1 - rank(d[],k)
        best[k]<- equals(rk[k],1)
}</pre>
```

```
}
```